

# Leveraging Free AI and ML Resources for Drug Optimization

Boutique R&D Services from Switzerland

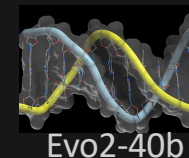
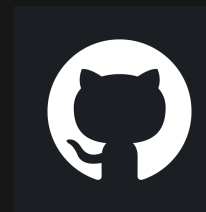
*Oliv Eidam*

CDD Community Meeting :: Feb 27, 2025



# Free AI / ML Resources Leveraging my Work

- AlphaFold2
- ML models and scripts on git
- Public datasets (ChEMBL and specialized sets)
- LLM models for biology (ESM-2, EVO-2)
- Antibody models (Ablang2)
- ChemProp (molecular representation)
- RDKit, DataWarrior (free tools for cheminfo)
- ChatGPT



Oxford Protein Informatics Group (OPIG)

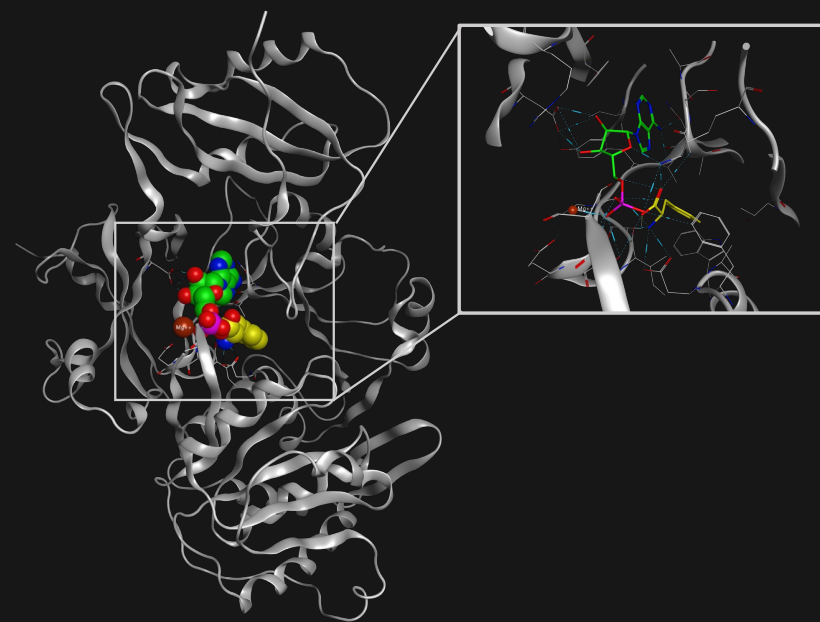


## Next: 3 Case Studies

Image sources: <https://alphafold.ebi.ac.uk>, <https://github.com>,  
<https://about.gitlab.com>, <https://www.rdkit.org>,  
<https://openmolecules.org/datawarrior>, <https://build.nvidia.com/arc/evo2-40b>,  
<https://opig.stats.ox.ac.uk>, <https://chatgpt.com>

# Study 1: AlphaFold2 for Homology Modeling

- **Challenge:** Model a target with 38% sequence identity
- **Benefits:** Fast, deep alignments, fewer human errors, available
- **Luck:** AF2 matched closed, inhibitor-bound form which was desired for inhibitor modeling
- **Guidance:** Used 2D images of unpublished X-rays
- **Limitation:** Co-factors, metals, inhibitors required separate modeling



*Structure depicts X-ray of enzyme with ADP (green), Mg<sup>2+</sup> (brown sphere) and a ligand (yellow). This is not the modeled target.*

# Study 2: ChemProp for Property Predictions

- **ChemProp:** Graph Neural Network (GNN) representation of molecules works well for property prediction of small molecules<sup>[1]</sup>
- **Peptide Membrane permeability :** ChemProp equally predictive (RMSE 0.75) and easier to use than alternative deep learning method (RMSE 0.77)<sup>[2]</sup>
- **Data and scripts:** see <sup>[3]</sup> and <sup>[4]</sup>

[1] Clemens Isert, Jimmy C. Kromann, Nikolaus Stiefl, Gisbert Schneider, Richard A. Lewis, ACS Omega. **2023** Jan 4;8(2):2046-2056. Machine Learning for Fast, Quantum Mechanics-Based Approximation of Drug Lipophilicity

[2] Jianan Li, Keisuke Yanagisawa, Yutaka Akiyama, Brief Bioinform. **2024** Jul 25;25(5). CycPeptMP: enhancing membrane permeability prediction of cyclic peptides with multi-level molecular features and data augmentation

[3] [https://github.com/cisert/rescross\\_logp\\_ml](https://github.com/cisert/rescross_logp_ml)

[4] CycPeptMPDB: <http://cycpeptmpdb.com>

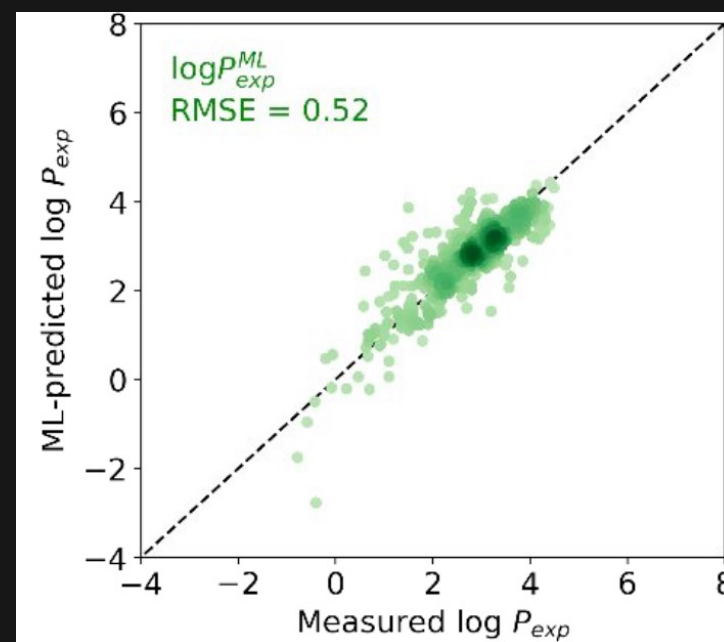


Figure S14 from [1] depicting ChemProp-predicted  $\log P$  against experimentally measured  $\log P$  for 425 test set molecules from public AZ dataset (2538 mols).



# Study 3: Antibody Modeling Boosted by AI

- **AF2-multimer:** Good for scaffolds, less confident in long CDRs<sup>[1]</sup>
- **AbLang2:** best freely available antibody sequence model<sup>[2]</sup>
- **ESM-2:** Boosts affinity, stability, reduces poly-specificity and immunogenicity<sup>[3,4]</sup>
- **Evo-2:** Largest bio-LLM<sup>[4]</sup>

[1] My favorite Jupyter notebook implementation of AF2:

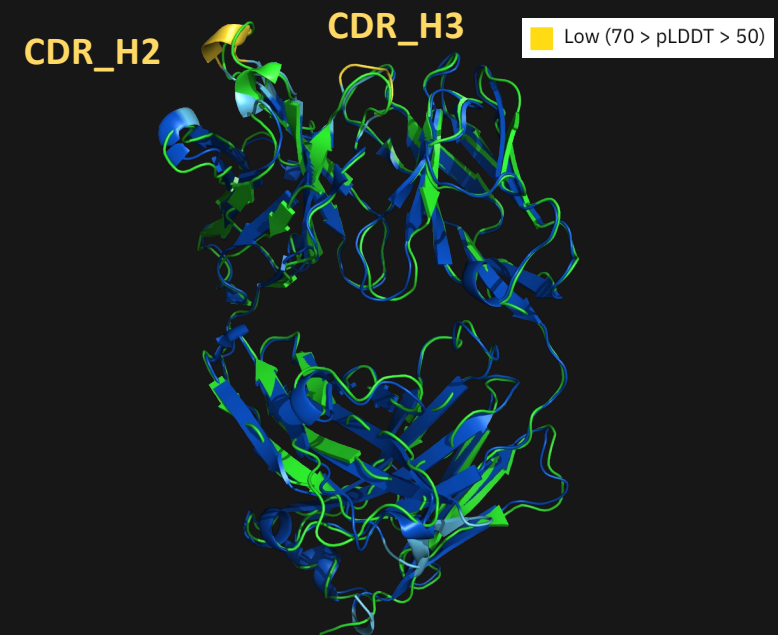
<https://colab.research.google.com/github/sokrypton/ColabFold/>

[2] <https://github.com/oxpig/AbLang2>

[3] Lin Z. et al, *Science*. **2023** Mar 17;379(6637):1123-1130. Evolutionary-scale prediction of atomic-level protein structure with a language model

[4] Brian Hie et al, *Nat Biotechnol*. **2024** Feb;42(2):275-283. Efficient evolution of human antibodies from general protein language models.

[5] Brian Hie and co-workers, *bioRxiv*, **Feb 19, 2025**. [DOI](#). Genome modeling and design across all domains of life with Evo 2



Superposition of AF2-model (blue) on Fab X-ray (green) shows little error (RMSD 1.16 Å) and perfect hetero-dimer of heavy and light chains. Only CDR\_H2 and CDR\_H3 exhibit low confidence (70 > pLDDT > 50).



# AI in Daily Use: Benefits & Limitations. Next?

## AI in Daily Use

- AI improves my emails and coding
- Study 1: AlphaFold2 reduces modeling errors, but ligands require separate modeling
- Study 2: ChemProp elevates peptide permeability prediction
- Study 3: AI models are powerful in antibody modeling but struggle with variable loops
- All Studies: Open-source code & data are valuable

## Next Steps: Let's Discuss!

- AlphaFold\*: Small molecule integration?
- New ML molecular representations?
- Merging physics with ML for predictions?

Thank you!

